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# REMARKS ON NONSMOOTH DYNAMIC VECTOR OPTIMIZATION PROBLEMS (Decision Theory in Mathematical Modelling)

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# REMARKS ON NONSMOOTH DYNAMIC VECTOR OPTIMIZATION PROBLEMS

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**1. Introduction.** This paper deals with vector optimization problems. By convention, throughout this paper we will use the following notations. For  $y = (y_1, \dots, y_n)$ ,  $z = (z_1, \dots, z_n) \in R^n$ , we say that

- (i)  $y \leq z$ , if and only if  $y_i \leq z_i$  for any  $i \in \{1, \dots, n\}$ ,
- (ii)  $y < z$  if and only if  $y_i \leq z_i$  for any  $i \in \{1, \dots, n\}$  with  $y \neq z$ ,
- (ii)  $y \ll z$  if and only if  $y_i < z_i$  for any  $i \in \{1, \dots, n\}$ .

Recently, many papers have been devoted to optimality conditions for the vector-valued programming and optimal control problems under some smooth or convex assumptions (see [2], [6], [7], [9], [10]). In [11], we derived the Kuhn-Tucker type proper-efficiency conditions for vector optimal control problems in general case. In this paper we use analogous method to discuss weak-efficiency and efficiency conditions for the following problem,

$$(P): \quad \begin{aligned} &\text{minimize: } \mathcal{F}(x, u) := (\mathcal{F}_1(x, u), \dots, \mathcal{F}_k(x, u)) \\ &\text{subject to: } \dot{x}(t) = \Phi(t, x(t), u(t)) \quad \text{a.e.}, \\ &\quad \quad \quad x(0) \in D, \quad u(t) \in U(t) \quad \text{a.e.}, \\ &\quad \quad \quad \mathcal{G}(x, u) := (\mathcal{G}_1(x, u), \dots, \mathcal{G}_l(x, u)) \leq 0 \end{aligned}$$

where

$$\begin{aligned} \mathcal{F}_i(x, u) &:= \int_0^1 F_i(t, x(t), u(t))dt + f_i(x(1)) \text{ for } i \in I := \{1, \dots, k\} \\ \mathcal{G}_j(x, u) &:= \int_0^1 G_j(t, x(t), u(t))dt + g_j(x(1)) \text{ for } j \in J := \{1, \dots, l\}; \end{aligned}$$

$x(\cdot) \in AC([0, 1], R^m)$  and  $u(\cdot) \in M([0, 1], R^n)$ ;  $F_i, G_j : [0, 1] \times R^m \times R^n \rightarrow R$ ,  $f_i, g_j : R^m \rightarrow R$  for  $i \in I, j \in J$  and  $\Phi : [0, 1] \times R^m \times R^n \rightarrow R^m$  are given functions;  $D$  is a subset of  $R^m$  and  $U(\cdot) : [0, 1] \rightarrow 2^{R^n}$  is a set-valued function. Here,  $AC([0, 1], R^m)$  is the space of absolutely continuous functions on  $[0, 1]$  with value in  $R^m$ ,  $M([0, 1], R^n)$  is the space of Lebesgue measurable functions on  $[0, 1]$  with value in  $R^n$ .

For this optimal control problem  $(P)$ , we say that  $(x, u)$  is an admissible process iff  $F_i(\cdot, x(\cdot), u(\cdot))$  and  $G_j(\cdot, x(\cdot), u(\cdot))$  are integrable for every  $i \in I$  and  $j \in J$ ,  $(x, u)$  satisfies state equation  $\dot{x}(t) = \Phi(t, x(t), u(t))$  a.e. with  $x(0) \in D$ ,  $u(t) \in U(t)$  a.e. and  $\mathcal{G}(x, u) \leq 0$ . The first component of a process  $(x, u)$  is called a state and the second is called a control. We denote by  $\Omega$  the set of all admissible processes of  $(P)$ . The optimal solutions for  $(P)$  are defined in the following meaning.

*Definition 1:*  $(x_*, u_*) \in \Omega$  is said to be

- (i) a weakly-efficient solution for  $(P)$  if there exists no  $(x, u) \in \Omega$  such that

$$\mathcal{F}(x, u) \ll \mathcal{F}(x_*, u_*);$$

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(ii) an efficient solution for (P) if there exists no  $(x, u) \in \Omega$  such that

$$\mathcal{F}(x, u) < \mathcal{F}(x_*, u_*).$$

**Definition 2:**  $(x_*, u_*) \in \Omega$  is called a local weakly-efficient solution of type (I) (resp. (II)) for (P) if and only if there is no  $(x, u) \in \Omega$  with  $\|x - x_*\|_{L^\infty} \leq \epsilon$  for some  $\epsilon > 0$  (resp. with  $x(t) \in x_*(t) + \epsilon B_m$  and  $u(t) \in u_*(t) + \epsilon B_n$  for some  $\epsilon > 0$ , where  $B^m$  and  $B^n$  are unit closed balls of  $R^m$  and  $R^n$ , respectively) such that  $\mathcal{F}(x, u) \ll \mathcal{F}(x_*, u_*)$ .

The main method to obtain optimality conditions for multiobjective optimization problems is based on a replacement of the multiobjective problems by single-objective (scalar) optimization problems. The following results give the relationship between (P) and scalar optimization problems.

**Lemma 1:**  $(x_*, u_*) \in \Omega$  is a weakly-efficient (local weakly-efficient) solution of (P) if and only if  $(x_*, u_*)$  is an optimal (local optimal) solution of the following scalar optimization problem,

$$\begin{aligned} \min : & \max_{i \in I} (\mathcal{F}_i(x, u) - \mathcal{F}_i(x_*, u_*)) \\ \text{s. t. : } & (x, u) \in \Omega. \end{aligned}$$

*Proof.* By the definitions, it is easy to see that  $(x_*, u_*)$  is a weakly efficient of (P) if and only if there is no  $(x, u) \in \Omega$  satisfying

$$\max_{i \in I} (\mathcal{F}_i(x, u) - \mathcal{F}_i(x_*, u_*)) < 0.$$

Thus, this lemma hold.  $\square$

**Lemma 2:** ([6, Lemma 3.1])  $(x_*, u_*) \in \Omega$  is an efficient solution of (P) if and only if  $(x_*, u_*)$  is an optimal solution of the following scalar optimal control problem  $(P_i)$  for each  $i \in I$ .

$$\begin{aligned} (P_i) : \quad & \text{minimize : } \mathcal{F}_i(x, u) \\ & \text{subject to : } (x, u) \in \Omega \\ & \mathcal{F}_j(x, u) - \mathcal{F}_j(x_*, u_*) \leq 0 \quad j \in I/\{i\}. \end{aligned}$$

**Lemma 3:** Suppose that  $\Omega$  is convex set and  $\mathcal{F}_i(x, u)$ ,  $i = 1, \dots, k$  are convex functions. Then,  $(x_*, u_*) \in \Omega$  is a weakly-efficient solution of (P) if and only if  $(x_*, u_*)$  is an optimal solution of  $(P_i)$  stated in Lemma 2 for some  $i \in I$ .

*Proof.* Assume that  $(x_*, u_*)$  is a weakly-efficient solution of (P). If for every  $(P_i)$ ,  $(x_*, u_*)$  is not an optimal solution, i.e. for any  $i \in I$  there exists  $(x_i, u_i) \in \Omega$  with

$$\begin{aligned} \mathcal{F}_i(x_i, u_i) &< \mathcal{F}_i(x_*, u_*) \\ \mathcal{F}_j(x_i, u_i) - \mathcal{F}_j(x_*, u_*) &\leq 0 \quad \text{for } j \in I/\{i\}. \end{aligned}$$

Putting  $(x_0, u_0) := \frac{1}{k} \sum_{i \in I} (x_i, u_i)$ , we see that  $(x_0, u_0) \in \Omega$ . Notice that  $\mathcal{F}_i(x, u)$  is convex, we have

$$\mathcal{F}_i(x_0, u_0) \leq \sum_{j \in I} \frac{1}{k} \mathcal{F}_i(x_j, u_j) < \mathcal{F}_i(x_*, u_*).$$

Thus,  $\mathcal{F}(x_0, u_0) \ll \mathcal{F}(x_*, u_*)$ , which contradicts that  $(x_*, u_*)$  is a weakly-efficient solution of  $(P)$ .

Conversely, let  $(x_*, u_*)$  be an optimal solution of  $(P_i)$  for some  $i \in I$ . If  $(x_*, u_*)$  is not a weakly-efficient solution of  $(P)$ , then there is  $(x, u) \in \Omega$  satisfying

$$\mathcal{F}_i(x, u) < \mathcal{F}_i(x_*, u_*) \text{ and } \mathcal{F}_j(x, u) - \mathcal{F}_j(x_*, u_*) < 0 \text{ for } j \in I/\{i\},$$

which contradicts that  $(x_*, u_*)$  is an optimal solution of  $(P_i)$ .  $\square$

**2. Optimality conditions.** For simplicity, throughout this section we omit the variable  $t$  when it does not cause confusion, and abbreviate the arguments  $(t, x_*(t), u_*(t))$  to  $[t]$ , for instance, we write  $G_i[t] = G_i(t, x_*(t), u_*(t))$ . In Theorem 1 and 2 below, the notations  $\partial$  denote the Clarke generalized gradients and  $N_D, N_{U(t)}$  indicate the Clarke normal cones, while in Theorem 3 and 4, these notations stand for the subdifferentials and the normal cones in the sense of convex analysis, respectively.

The following assumptions are required. The pair  $(x_*, u_*)$  in (A2) and (A3) will be assumed to be a local weakly efficient solution of type (I) for  $(P)$ .

(A1):  $D$  is closed,  $U(\cdot)$  is a nonempty compact set-valued map and the graph  $GrU$  is  $\mathcal{L} \times \mathcal{B}$  measurable.

(A2):  $f_i(\cdot), g_j(\cdot)$  ( $i \in I, j \in J$ ) are Lipschitz continuous in a neighborhood of  $x_*(1) \in R^m$ .

(A3): For every admissible control  $u(\cdot)$ , there are real-valued measurable function  $\epsilon(t) > 0$  and  $h_i(t) \geq 0, i = 0, \dots, k+l$ , such that

$$\begin{aligned} |F_i(t, x, u(t)) - F_i(t, x', u(t))| &\leq h_i(t) |x - x'| \text{ for } i \in I \\ |G_j(t, x, u(t)) - G_j(t, x', u(t))| &\leq h_{k+j}(t) |x - x'| \text{ for } j \in J \\ |\Phi(t, x, u(t)) - \Phi(t, x', u(t))| &\leq h_0(t) |x - x'| \end{aligned}$$

whenever  $|x - x_*(t)| \leq \epsilon(t), |x' - x_*(t)| \leq \epsilon(t), t \in [0, 1]$ ; for  $u(\cdot) = u_*(\cdot)$  these functions can be chosen in such a way that  $\epsilon(t) = \epsilon > 0$  and  $h_i(t)$  ( $i = 0, \dots, k+l$ ) are integrable.

(A4): For any  $u(\cdot) \in \mathcal{U} := \{u(\cdot) \in M([0, 1], R^n) : u(t) \in U(t) \text{ a.e.}\}$ ,  $F_i(t, x, u(t))$  for  $i \in I, G_j(t, x, u(t))$  for  $j \in J$  and  $\Phi(t, x, u(t))$  are measurable.

**Theorem 1.** *Let assumptions (A1)-(A4) be satisfied. Suppose that  $(x_*, u_*)$  is a local weakly efficient solution of type (I) for  $(P)$ . Then, there exist  $\lambda = (\lambda_1, \dots, \lambda_{k+l}) > 0$  and an absolutely continuous function  $p(\cdot) : [0, 1] \rightarrow R^n$ , such that*

$$(1) \quad -\dot{p}(t) \in \partial_x H(t, x_*(t), p(t), u_*(t), \lambda) \quad \text{a.e.}$$

$$(2) \quad p(0) \in N_D(x_*(0)), -p(1) \in \sum_{i \in I} \lambda_i \partial f_i(x_*(1)) + \sum_{j \in J} \lambda_{k+j} \partial g_j(x_*(1))$$

$$(3) \quad H(t, x_*(t), p(t), u_*(t), \lambda) = \max_{v \in U(t)} H(t, x_*(t), p(t), v, \lambda) \quad \text{a.e.}$$

$$(4) \quad \lambda_{k+j} \left( \int_0^1 G_j[t] dt + g_j(x_*(1)) \right) = 0 \quad \text{for } j \in J$$

where  $H(t, x, p, u, \lambda) := \langle p, \Phi(t, x, u) \rangle - \sum_{i \in I} \lambda_i F_i(t, x, u) - \sum_{j \in J} \lambda_{k+j} G_j(t, x, u)$

*Proof.* We consider the following problem,

$$(P') \quad \begin{aligned} \min : \quad & \Gamma_0(y) := \max_{i \in I} \{y_i(1) + f_i(x(1)) - \mathcal{F}_i(x_*, u_*)\} \\ \text{s. t. :} \quad & L_0(y, u) := x(t) - x(0) - \int_0^t \Phi(t, x(t), u(t)) dt = 0 \\ & L_i(y, u) := y_i(t) - \int_0^t F_i(t, x(t), u(t)) dt = 0 \quad i \in I \\ & L_{k+j}(y, u) := y_{k+j}(t) - \int_0^t G_j(t, x(t), u(t)) dt = 0 \quad j \in J \\ & \Gamma_j(y) := y_{k+j}(1) + g_j(x(1)) \leq 0 \quad j \in J \\ & y(\cdot) \in \mathcal{S}, \quad u(\cdot) \in \mathcal{U}, \end{aligned}$$

where  $y(\cdot) := (x(\cdot), y_1(\cdot), \dots, y_{k+l}(\cdot)) \in C([0, 1], R^{m+k+l})$  is the state and  $u(\cdot) \in M([0, 1], R^n)$  is the control,  $\mathcal{S} := \{x \in C([0, 1], R^m) : x(0) \in D\} \times C([0, 1], R^{2k})$ .

Let  $y_{i*}(t) := \int_0^t F_i[t] dt$  for  $i \in I$  and  $y_{(k+j)*}(t) := \int_0^t G_j[t] dt$  for  $j \in J$ . Thus, by Lemma 1, we see that  $y_* := (x_*, y_{i*}, \dots, y_{(k+l)*})$  corresponding  $u_*$  minimizes  $\Gamma_0(y)$  over all admissible processes  $(y, u)$  for  $(P')$  with  $x$  being sufficiently close to  $x_*$  in the norm of  $L^\infty$ .

By [4, Theorem 2], we see that there exist Lagrange multipliers  $\delta := (\delta_0, \dots, \delta_l) \geq 0$ ,  $x^* \in C^*([0, 1], R^m)$ , and  $y_i^* \in C^*([0, 1], R)$   $i = 1, \dots, k+l$  not all zero such that

$$(5) \quad 0 \in \partial_y \mathcal{L}(y_*, y^*, u_*, \kappa) + N_{\mathcal{S}}(y_*)$$

$$(6) \quad \mathcal{L}(y_*, y^*, u_*, \kappa) = \min_{u \in \mathcal{U}} \mathcal{L}(y_*, y^*, u, \kappa)$$

$$(7) \quad \delta_j \Gamma_j(y_*) = 0 \quad j \in J$$

where  $\mathcal{L}(y, y^*, u, \kappa) := \sum_{i=0}^l \delta_i \Gamma_i(y) + \langle x^*, L_0(y, u) \rangle + \sum_{i=1}^{k+l} \langle y_i^*, L_i(y, u) \rangle$ .

According to the formulas of the Clarke gradients (see [3]), we see that

(i) For any  $\xi \in \partial \Gamma_0(y_*)$ , there are  $\bar{\lambda}_i \geq 0$ ,  $\nu_i \in \partial f_i(x_*(1))$  for  $i \in I$  with  $\sum_{i \in I} \bar{\lambda}_i = 1$  such that for any  $y \in C([0, 1], R^{n+2k})$

$$\langle \xi, y \rangle = \sum_{i \in I} \bar{\lambda}_i y_i(1) + \sum_{i \in I} \bar{\lambda}_i \langle \nu_i, x(1) \rangle.$$

for every  $\xi \in \sum_{i=1}^l \delta_i \Gamma_i(y_*)$ , there exist  $\nu_{k+j} \in \partial g_j(x_*(1))$  for  $j \in J$  such that for any  $y \in C([0, 1], R^{n+2k})$

$$\langle \xi, y \rangle = \sum_{j \in J} \delta_j y_i(1) + \sum_{j \in J} \delta_j \langle \nu_{k+j}, x(1) \rangle.$$

Analyzing as in [4], we have the following.

(ii) The above multipliers  $x^*, y_1^*, \dots, y_{2k}^*$  can be expressed by pairs of the nonnegative Radon measure and Radon-integrable functions  $(\mu_i, \xi_i)$ ,  $i = 0, \dots, 2k$ . For every  $\xi \in \partial_y (\langle x^*, L_0(x_*, u_*) \rangle + \sum_{i=1}^{k+l} \langle y_i^*, L_i(x_*, u_*) \rangle)$ , there is a Lebesgue measurable function  $\eta(\cdot)$  with

$$(8) \quad \begin{aligned} \eta(t) \in \quad & \partial_x \left( \left\langle \int_t^1 \xi_0 d\mu_0, \Phi[t] \right\rangle + \sum_{i \in I} \left\langle \int_t^1 \xi_i d\mu_i, F_i(t, x_*(t), u_*(t)) \right\rangle \right. \\ & \left. + \sum_{j \in J} \left\langle \int_t^1 \xi_{k+i} d\mu_{k+i}, G_i(t, x_*(t), u_*(t)) \right\rangle \right) \quad \text{a.e.}, \end{aligned}$$

such that for any  $y \in C([0, 1], R^{n+2k})$ ,

$$\langle \xi, y \rangle = \int_0^1 \langle x(t) - x(0), \xi_0 \rangle d\mu_0 + \sum_{i=1}^{k+l} \int_0^1 \langle y_i, \xi_i \rangle d\mu_i - \int_0^1 \langle \eta, x \rangle dt.$$

(iii) For each  $\xi \in N_S(y_*)$ , there is  $\alpha \in N_D(x_*(0))$ , such that

$$\langle \xi, y \rangle := \langle \alpha, x(0) \rangle \quad \text{for any } y \in C([0, 1], R^{n+k}).$$

Combining (i), (ii) and (iii), from (5) we see that there are  $\bar{\lambda}_i$ ,  $i = 1, \dots, l$ ;  $\nu_i$ ,  $i = 1, \dots, k+l$ ;  $(\mu_i, \xi_i)$ ,  $i = 0, \dots, k+l$ ,  $\eta$  and  $\alpha$  stated above such that

$$\begin{aligned} 0 = & \sum_{i \in I} \delta_0 \bar{\lambda}_i y_i(1) + \sum_{j \in J} \delta_j y_{k+j}(1) + \sum_{i \in I} \delta_0 \bar{\lambda}_i \langle \nu_i, x(1) \rangle + \sum_{j \in J} \delta_j \langle \nu_{k+j}, x(1) \rangle + \\ & \sum_{i=1}^{k+l} \int_0^1 \langle y_i, \xi_i \rangle d\mu_i + \int_0^1 \langle x(t) - x(0), \xi_0 \rangle d\mu_0 - \int_0^1 \langle \eta, x \rangle dt + \langle \alpha, x(0) \rangle \end{aligned}$$

for any  $x \in C([0, 1], R^n)$  and  $y_i \in C([0, 1], R)$ ,  $i = 1, \dots, k+l$ .

Setting  $\lambda_i = \delta_0 \bar{\lambda}_i$  for  $i \in I$ ,  $\lambda_{k+j} := \delta_j$  for  $j \in J$  and  $p(t) := \int_t^1 \xi_0 d\mu_0$ , from the above equation, we see that

$$\begin{aligned} \lambda_i y_i(1) + \int_0^1 \left\langle \int_t^1 \xi_i d\mu_i, \dot{y}_i \right\rangle dt &= 0 \quad (\forall y_i \in AC \text{ with } y_i(0) = 0, i \in I \cup J), \\ \langle \alpha, x(0) \rangle + \sum_{i=1}^{k+l} \lambda_i \langle \nu_i, x(1) \rangle + \int_0^1 \left\langle p(t) - \int_t^1 \eta d\tau, \dot{x} \right\rangle dt &= 0 \quad (\forall x \in AC). \end{aligned}$$

These yield that (refer to the proof of [4, Theorem 3])

$$\begin{aligned} (9) \quad & \int_t^1 \xi_i d\mu_i = -\lambda_i, \quad i = 1, \dots, k+l \\ & \dot{p}(t) = -\eta(t) \text{ a.e.}, \quad p(0) = \alpha, \quad p(1) = -\sum_{i=1}^{k+l} \lambda_i \nu_i. \end{aligned}$$

Therefore, (9), (8) and (7) imply (1), (2) and (4)

Here, if  $\delta = 0$ , then  $(\lambda_1, \dots, \lambda_{k+l}) = (y_1^*, \dots, y_{k+l}^*) = 0$ . From (1) and (2), we can get  $p(\cdot) = 0$ . Thus,  $y^* = 0$  which contradicts that  $\delta$  and  $y^*$  are not all zero. Hence, we have  $(\lambda_1, \dots, \lambda_{k+l}) > 0$ .

On other hand, By (6) and (9), we see that

$$\int_0^1 H(t, x_*, p, u_*, \lambda) dt = \max_{u \in \mathcal{U}} \int_0^1 H(t, x_*, p, u, \lambda) dt.$$

Discussing as in the proof of [4, Theorem 3], we can obtain (3).  $\square$

According to the results of [8], we see that the above necessary conditions (1)-(4) (Maximum Principle-type) may fail to be sufficient conditions for weak-efficient solutions of (P) even in the “convex” case given below. Next, we give another type necessary weakly-efficiency conditions for (P), which is an extension of [8]. In the “convex” case, the latter necessary conditions are necessary-sufficient for weakly-efficiency under Slater constraint qualifications. Moreover, these conditions are also necessary-sufficient for efficient solutions of (P) under further assumptions.

We impose the following assumption, in which the process  $(x_*, u_*)$  will be assumed to be a weakly-efficient solution of type (II) for (P).

(A5):  $F_i(\cdot, x, u)$ ,  $G_i(\cdot, x, u)$ ,  $i = 1, \dots, k$ ,  $\Phi(\cdot, x, u)$  are Lebesgue measurable, and there exist  $\epsilon > 0$  and  $h_i(t) \in L^1([0, 1], R)$ ,  $i = 0, \dots, k+l$ , such that

$$\begin{aligned} |F_i(t, x, u) - F_i(t, x', u')| &\leq h_i(t) (|x - x'| + |u - u'|) \quad \text{for } i \in I \\ |G_j(t, x, u) - G_j(t, x', u')| &\leq h_{k+j}(t) (|x - x'| + |u - u'|) \quad \text{for } j \in J \\ |\Phi(t, x, u(t)) - \Phi(t, x', u'(t))| &\leq h_0(t) (|x - x'| + |u - u'|) \end{aligned}$$

whenever  $x, x' \in x_*(t) + \epsilon B_n$ ,  $u, u' \in u_*(t) + \epsilon B_m$  a.e..

**Theorem 2:** Assume that (A1), (A2) and (A5) be satisfied. Let  $(x_*, u_*)$  be a local weakly efficient solution of type (II) for (P). Then there exist  $\lambda = (\lambda_1 \dots, \lambda_{k+l}) > 0$ , an absolutely continuous function  $p(\cdot) : [0, 1] \rightarrow R^n$  and an integrable function  $\zeta(\cdot) : [0, 1] \rightarrow R^m$  such that

$$(10) \quad (-\dot{p}(t), \zeta(t)) \in \partial_{(x,u)} H(t, x_*(t), p(t), u_*(t), \lambda) \quad a.e.$$

$$(11) \quad p(0) \in N_D(x_*(0)), \quad -p(1) \in \sum_{i \in I} \lambda_i \partial f_i(x_*(1)) + \sum_{j \in J} \lambda_{k+j} \partial g_j(x_*(1))$$

$$(12) \quad \zeta(t) \in N_{U(t)}(u_*(t)) \quad a.e.$$

$$(13) \quad \lambda_{k+j} \left( \int_0^1 G_j[t] dt + g_j(x_*(1)) \right) = 0 \quad \text{for } j \in J$$

where  $H(t, x, p, u, \lambda)$  is defined in Theorem 1.

*Proof.* It is obvious that the scalar optimization problem in Lemma 1 can be rewritten as follows

$$\begin{aligned} (P^\dagger) : \quad & \text{minimize : } \Gamma(y(1)) := \max_{i \in I, j \in J} \{y_i(1) + f_i(x(1)) - \mathcal{F}_i(x_*, u_*) \\ & \quad \quad \quad y_{k+j}(1) + g_j(x(1))\} \\ & \text{subject to : } \dot{x}(t) = \Phi(t, x(t), u(t)) \quad a.e. \\ & \quad \quad \dot{y}_i(t) = F_i(t, x(t), u(t)) \quad a.e. \quad i \in I \\ & \quad \quad \dot{y}_{k+i}(t) = G_i(t, x(t), u(t)) \quad a.e. \quad i \in I \\ & \quad \quad x(0) \in C, \quad y_i(0) = 0 \quad i = 1, \dots, 2k, \\ & \quad \quad u(t) \in U(t) \quad a.e. \end{aligned}$$

where  $y := (x, y_1, \dots, y_{2k}) \in AC([0, 1], R^{m+2k})$  is the state and  $u \in M([0, 1], R^n)$  is the control.

Define  $y_*$  as in proof of Theorem 1. By Lemma 1, we see that  $(y_*, u_*)$  is a minimizer over all admissible process for  $(P^\dagger)$  with  $x(t) \in x_*(t) + \epsilon B_n$ ,  $u(t) \in u_*(t) + \epsilon B_m$  a.e. for some  $\epsilon > 0$ . Thus, by [8, Proposition 6.1], there exist an absolutely continuous function  $\bar{p} = (p, p_1, \dots, p_{k+l})$  and an integrable function  $\zeta$  such that (12) and the following hold

$$(14) \quad (-\dot{\bar{p}}(t), \dot{y}(t), \zeta(t)) \in \partial_{(y, \bar{p}, u)} \bar{H}(t, y_*(t), \bar{p}(t), u_*(t)) \quad a.e.$$

$$(15) \quad \bar{p}(0) \in N_{C \times \underbrace{\{0\} \times \dots \times \{0\}}_{2k}}(y_*(0))$$

$$(16) \quad -\bar{p}(1) \in \partial\Gamma(y_*(1))$$

where  $\bar{H}(t, y, \bar{p}, u) := \langle p, \Phi(t, x, u) \rangle + \sum_{i \in I} \langle p_i, F_i(t, x, u) \rangle + \sum_{i \in I} \langle p_{k+i}, G_i(t, x, u) \rangle$ .

First, let us discuss inclusion (16). Notice that for every  $i \in I$  and  $j \in J$ ,

$$\begin{aligned} \Gamma_i(y(1)) &:= y_i(1) + f_i(x(1)) - \mathcal{F}_i(x_*, u_*), \\ \Gamma_j(y(1)) &:= y_{k+j}(1) + g_j(x(1)) \end{aligned}$$

only contains the arguments  $x$  and  $y_i$ , and  $\Gamma_i(y_*(1)) = \Gamma(y_*(1)) = 0$ . So by the formulas of the Clarke gradients, there are  $\gamma_i \in \partial_x f_i(x_*(1))$  for  $i \in I$ ,  $\gamma_{k+j} \in \partial_x g_j(x_*(1))$  for  $j \in J$  and  $(\lambda_1, \dots, \lambda_{k+l}) > 0$  such that

$$(17) \quad -p(1) = \sum_{i \in I} \lambda_i \gamma_i, \quad -p_i(1) = \lambda_i, \quad i = 1, \dots, k+l.$$

where we can set  $\lambda_j = 0$  for  $j \in \{j \in J : \mathcal{G}_j(x_*, u_*) < 0\}$ .

Thus, (11) and (13) follow from (15) and (17).

On the other hand, since  $\bar{H}$  does not contain the arguments  $y_i$ ,  $i = 1, \dots, k+l$ , (14) implies that  $\dot{p}_i(\cdot) = 0$ ,  $i = 1, \dots, k+l$ . Thus,  $p_i(\cdot) = -\lambda_i$ ,  $i = 1, \dots, k+l$  and

$$(-\dot{p}(t), \dot{x}(t), \zeta(t)) \in \partial_{(x, \bar{p}, u)} \left( \langle p(t), \Phi[t] \rangle - \sum_{i \in I} \lambda_i F_i[t] - \sum_{i \in I} \lambda_{k+i} G_i[t] \right) \quad a.e.$$

From this inclusion, by the definition of the Clarke generalized gradients, we can easily deduce (10).

Next, we proceed to the optimality conditions for the following problem.

$$\begin{aligned} (P^*) : \quad & \min : \mathcal{F}(x, u) \\ & s. t. : \dot{x}(t) = A(t)x(t) + B(t)u(t) + b(t) \quad a.e. \\ & \quad \quad x(0) \in D, \quad u(t) \in U(t) \quad a.e. \\ & \quad \quad \mathcal{G}(x, u) \leq 0 \end{aligned}$$

where  $x(\cdot) \in AC([0, 1], R^m)$  and  $u(\cdot) \in L^1([0, 1], R^n)$ ,  $\mathcal{F}$  and  $\mathcal{G}$  are given above,  $A(\cdot) : [0, 1] \rightarrow R^{n \times n}$ ,  $B(\cdot) : [0, 1] \rightarrow R^{n \times m}$  are integrable,  $b(\cdot) : [0, 1] \rightarrow R^n$  is measurable.

We impose the following hypotheses:

(H1): For every  $i \in I$ ,  $F_i(\cdot, x(\cdot), u(\cdot))$  and  $G_i(\cdot, x(\cdot), u(\cdot))$  are integrable for any  $(x, u) \in AC \times L^1$ .

(H2):  $F_i(t, \cdot, \cdot)$  for  $i \in I$  and  $G_i(t, \cdot, \cdot)$  for  $j \in J$  are convex lower semicontinuous, and there are  $v_i(t) \in L^\infty([0, 1], R^{m+n})$  and  $w_i(t) \in L^1([0, 1], R)$ ,  $i = 1, \dots, k+l$  such that for any  $x \in R^m$ ,  $u \in R^n$ ,  $F_i(t, x, u) \geq \langle v_i(t), (x, u) \rangle + w_i(t)$  for  $i \in I$  and  $G_j(t, x, u) \geq \langle v_j(t), (x, u) \rangle + w_j(t)$  for  $j \in J$  a.e..

(H3): The functions  $f_i(\cdot)$  for  $i \in I$  and  $g_i(\cdot)$  for  $j \in J$  are proper convex and lower semicontinuous.

(H4): The set  $C$  is convex,  $U(t)$  is convex a.e., and there is  $\rho(t) \in L^1$  such that  $|u| \leq \rho(t)$  for any  $u \in U(t)$  a.e..

(H5): There exists an admissible process  $(x_i, u_i)$  for  $(P^*)$ , such that  $\mathcal{G}_j(x_i, u_i) - \mathcal{G}_j(x_*, u_*) < 0$  for any  $j \in \{j \in J : \mathcal{G}_j(x_*, u_*) = 0\}$ .

Here,  $(x_*, u_*)$  will be assumed to be an admissible process for  $(P^*)$ .



**Theorem 3:** Assume that (H1)-(H5) and (A1) be satisfied. An admissible process  $(x_*, u_*)$  is a weakly-efficient solution for  $(P^*)$  if and only if there exist  $\lambda = (\lambda_1, \dots, \lambda_{k+l}) \geq 0$  with  $(\lambda_1, \dots, \lambda_k) > 0$ ,  $p(\cdot) \in AC([0, 1], R^m)$ , and  $\zeta(\cdot) \in L^\infty([0, 1], R^n)$  such that

$$(18) \quad (\dot{p}(t) + p(t)A(t), p(t)B(t) - \zeta(t)) \in \partial_{(x,u)} \left( \sum_{i \in I} \lambda_i F_i[t] + \sum_{j \in J} \lambda_{k+j} G_j[t] \right) \quad a.e.$$

$$(19) \quad p(0) \in N_C(x_*(1)), \quad -p(1) \in \sum_{i \in I} \lambda_i \partial f_i(x_*(1)) + \sum_{j \in J} \lambda_{k+j} \partial g_j(x_*(1))$$

$$(20) \quad \zeta(t) \in N_{U(t)}(u_*(t)) \quad a.e.,$$

$$(21) \quad \lambda_{k+j} \left( \int_0^1 G_j[t] dt + g_j(x_*(1)) \right) = 0 \quad \text{for } j \in J.$$

*Proof.* [Necessity] By Lemma 3, we know that there exists  $i \in I$  such that  $(x_*, u_*)$  is an optimal solution for the following scalar optimal control problem,

$$\begin{aligned} & \text{minimize : } \mathcal{F}_i(x, u) \\ & \text{subject to : } \dot{x}(t) - A(t)x(t) - B(t)u(t) - b(t) = 0 \quad a.e. \\ & \quad \mathcal{G}_j(x, u) \leq 0 \quad j \in J \\ & \quad \mathcal{F}_j(x, u) \leq 0 \quad j \in I/\{i\} \\ & \quad x \in \{x \in AC([0, 1], R^m) : x(0) \in D\} \\ & \quad u \in \mathcal{C} := \{u \in L^1([0, 1], R^n) : u(t) \in U(t) \text{ a.e.}\}. \end{aligned}$$

This means that  $(x_*, u_*, x_*(0), x_*(1))$  is a minimizer for the following scalar optimization problem.

$$\begin{aligned} & \text{minimize : } \Lambda_i(z, u, \alpha, \beta) := \int_0^1 F_i(t, z, u) dt + f_i(\beta) \\ & \text{subject to : } \Gamma_1(z, u, \alpha, \beta) := z(t) - \alpha - \int_0^t (Az + Bu + b) d\tau = 0 \quad a.e. \\ & \quad \Gamma_2(z, u, \alpha, \beta) := \beta - \alpha - \int_0^1 (Az + Bu + b) d\tau = 0 \\ & \quad \Lambda_j(z, u, \alpha, \beta) := \int_0^1 F_j(t, z, u) dt + f_j(\beta) - \mathcal{F}_j(x_*, u_*) \leq 0 \text{ for } j \in I/\{i\} \\ & \quad \Lambda_j(z, u, \alpha, \beta) := \int_0^1 G_j(t, z, u) dt + g_j(\beta) \leq 0 \text{ for } j \in J \\ & \quad (z, u, \alpha, \beta) \in \mathcal{M} := L^1([0, 1], R^m) \times \mathcal{C} \times D \times R^m, \end{aligned}$$

where  $(z, u, \alpha, \beta) \in L^1([0, 1], R^m) \times L^1([0, 1], R^n) \times R^m \times R^m$

Put  $\theta := (z, u, \alpha, \beta)$  and  $\theta_* := (x_*, u_*, x_*(0), x_*(1))$ . It is obvious that  $\Lambda_i(\theta)$  is convex,  $\Gamma_1(\theta)$  and  $\Gamma_2(\theta)$  are affine mappings. By [5, Theorem 5 p74], there exist  $\lambda := (\lambda_1, \dots, \lambda_{k+l}) \geq 0$ ,  $q(\cdot) \in (L^1)^*$  and  $\sigma \in R^m$  not all zero, such that

$$(22) \quad \begin{aligned} & \sum_{j=1}^{k+l} \lambda_j \Lambda_j(\theta_*) + \int_0^1 \langle q, \Gamma_1(\theta_*) \rangle dt + \langle \sigma, \Gamma_2(\theta_*) \rangle \\ & = \min_{\theta \in \mathcal{M}} \left( \sum_{j=1}^{k+l} \lambda_j \Lambda_j(\theta) + \int_0^1 \langle q, \Gamma_1(\theta) \rangle dt + \langle \sigma, \Gamma_2(\theta) \rangle \right), \end{aligned}$$

$$\lambda_{k+j}\Lambda_j(\theta_*) = \lambda_{k+j} \left( \int_0^1 G_j[t]dt + g_j(x_*(1)) \right) = 0 \quad \text{for } j \in J$$

Let  $I_{\mathcal{M}}(\theta)$  denote the indicator function of  $\mathcal{M}$ . Notice that the functions  $I_{\mathcal{M}}$ ,  $\Lambda_j$  ( $j \in I$ ),  $\int_0^1 \langle p, \Gamma_1 \rangle dt$ ,  $\langle \sigma, \Gamma_2 \rangle$  are proper convex and lower semicontinuous, from (22) we see that

$$(23) \quad 0 \in \sum_{j=1}^{k+l} \lambda_j \partial \Lambda_j(\theta_*) + \partial \int_0^1 \langle q, \Gamma_1(\theta_*) \rangle dt + \partial \langle \sigma, \Gamma_2(\theta_*) \rangle + N_{\mathcal{M}}(\theta_*)$$

(refer to Section 1 of Chapter 1 in [1]).

Now, we analyze (23). By the formulas of subdifferential (see [1], [5]), we have the following conclusions.

For every  $\xi \in \sum_{j=1}^{k+l} \lambda_j \partial \Lambda_j(\theta_*)$ , there are  $(\mu_j, \eta_j) \in L^\infty$  with  $(\mu_j(t), \eta_j(t)) \in \partial_{(x,u)} F_i[t]$  and  $\nu_j \in \partial f_j(x_*(1))$  for  $j \in I$ ,  $(\mu_{k+j}, \eta_{k+j}) \in L^\infty$  with  $(\mu_{k+j}(t), \eta_{k+j}(t)) \in \partial_{(x,u)} G_i[t]$  and  $\nu_{k+j} \in \partial g_j(x_*(1))$  for  $j \in J$  such that for any  $\theta \in L^1 \times L^1 \times R^m \times R^m$

$$\langle \xi, \theta \rangle = \sum_{j=1}^{k+l} \lambda_j \left( \int_0^1 (\langle \mu_j, x \rangle + \langle \eta_j, u \rangle) dt + \langle \nu_j, \beta \rangle \right).$$

Corresponding to any  $\xi \in N_{\mathcal{M}}(\theta_*)$ , there are  $\gamma \in N_D(x_*(0))$ , and  $\zeta(\cdot) \in N_C(u_*(\cdot))$  such that for any  $\theta \in L^1 \times L^1 \times R^m \times R^m$ , one has

$$\langle \xi, \theta \rangle = \langle \gamma, \alpha \rangle + \int_0^1 \langle \zeta, u \rangle dt.$$

Notice that  $\int_0^1 \langle q, \Gamma_1(\theta) \rangle dt$  is affine on  $\theta$ , thus  $\partial \int_0^1 \langle q, \Gamma_1(\theta_*) \rangle dt = \{\xi\}$  with

$$\langle \xi, \theta \rangle = \int_0^1 \left\langle q, z - \alpha - \int_0^t (Az - Bu) d\tau \right\rangle dt$$

for any  $\theta \in L^1 \times L^1 \times R^m \times R^m$ .

Similarly,  $\partial \langle \sigma, \Gamma_2(\theta_*) \rangle = \{\xi\}$  with

$$\langle \xi, \theta \rangle = \left\langle \sigma, \beta - \alpha - \int_0^1 (Az - Bu) dt \right\rangle$$

for any  $\theta \in L^1 \times L^1 \times R^m \times R^m$ .

Then, (23) implies that there are  $(\mu_j, \eta_j)$ ,  $\nu_j$ ,  $j = 1, \dots, k+l$ ,  $\gamma$  and  $\zeta$  stated above such that

$$(24) \quad \sum_{j=1}^{k+l} \lambda_j \int_0^1 (\langle \mu_j, z \rangle + \langle \eta_j, u \rangle) dt + \sum_{j=1}^{k+l} \lambda_j \langle \nu_j, \beta \rangle + \int_0^1 \left\langle q, z - \int_0^t (Az + Bu) d\tau \right\rangle dt \\ - \left\langle \int_0^1 q dt, \alpha \right\rangle + \left\langle \sigma, \beta - \alpha - \int_0^1 (Az + Bu) dt \right\rangle + \langle \gamma, \alpha \rangle + \int_0^1 \langle \zeta, u \rangle dt = 0$$

for any  $(z, u, \alpha, \beta) \in L^1 \times L^1 \times R^m \times R^m$ .

Put  $p(t) := \int_t^1 q(\tau) d\tau + \sigma$ . From (24) we see that

$$\int_0^1 \left\langle \sum_{i=1}^{k+l} \lambda_i \mu_i, z \right\rangle dt - \int_0^1 \langle \dot{p} + pA, z \rangle dt + \int_0^1 \left\langle \sum_{i=1}^{k+l} \lambda_i \eta_i, u \right\rangle dt - \int_0^1 \langle pB - \zeta, u \rangle dt \\ + \left\langle \sum_{i=1}^{k+l} \lambda_i \nu_i, \beta \right\rangle + \langle \sigma, \beta \rangle - \left\langle \int_0^1 q dt, \alpha \right\rangle - \langle \sigma, \alpha \rangle + \langle \gamma, \alpha \rangle = 0$$

for any  $(z, u, \alpha, \beta) \in L^1 \times L^1 \times R^m \times R^n$ , which implies that

$$(25) \quad \begin{aligned} \dot{p} + pA &= \sum_{i=1}^{k+l} \lambda_i \mu_i, \quad pB - \zeta = \sum_{i=1}^{k+l} \lambda_i \eta_i, \\ p(1) &= \sigma = - \sum_{i=1}^{k+l} \lambda_j \nu_j, \quad p(0) = \int_0^1 q(\tau) d\tau + \sigma = \gamma. \end{aligned}$$

From (25), we obtain (18) and (19).

By  $\zeta(\cdot) \in N_C(u_*(\cdot))$ , we have  $\zeta(t)(u(t) - u_*(t)) \leq 0$  for any  $u(\cdot) \in \mathcal{U}$ . Thus, from the theory of measurable selection (20) follows.

Finally, if  $\lambda = 0$ , then (28) and (29) imply that  $\sigma = 0$  and  $p(\cdot) = 0$ , thus  $\lambda, q$  and  $\sigma$  all are zero. Hence,  $\lambda > 0$ . If  $(\lambda_1, \dots, \lambda_k) = 0$ , then  $(\lambda_k, \dots, \lambda_{k+l}) > 0$ . By the Slater constraint qualifications (H5) and the conditions (18)-(21), we have that

$$\begin{aligned} 0 &> \sum_{j \in J} \lambda_{k+j} (\mathcal{G}_j(x_i, u_i) - \mathcal{G}_j(x_*, u_*)) \\ &= \sum_{j \in I/\{i\}} \lambda_j \left( \int_0^1 (G_j(t, x_i, u_i) - G_j[t]) dt + g_j(x_i(1)) - g_j(x_*(1)) \right) \\ &\geq \int_0^1 (\langle \dot{p} + pA, x_i - x_* \rangle + \langle pB - \zeta, u_i - u_* \rangle) dt - p(1)(x_i(1) - x_*(1)) \\ &= -p(0)(x_i(0) - x_*(0)) - \int_0^1 \langle \zeta, u_i - u_* \rangle dt \\ &\geq 0, \end{aligned}$$

a contradiction. Hence,  $(\lambda_1, \dots, \lambda_k) > 0$ .

[Sufficiency] Assume that there exist  $(\lambda_1, \dots, \lambda_k) > 0$ ,  $p(\cdot) \in AC$ , and  $\zeta(\cdot) \in L^\infty$  satisfying (18)-(21). Notice that  $\sum_{i \in I} \lambda_i > 0$ , so we can set  $\sum_{i \in I} \lambda_i = 1$ . Let  $(x, u)$  be an arbitrary admissible process for  $(P^*)$ . Using (18)-(21) again, we see that

$$\begin{aligned} &\max \{ \mathcal{F}_i(x, u) - \mathcal{F}_i(x_*, u_*) : i \in I \} \\ &\geq \sum_{i \in I} \lambda_i \left( \int_0^1 F_i(t, x, u) dt + f_i(x(1)) - \int_0^1 F_i[t] dt - f_i(x_*(1)) \right) \\ &\quad \sum_{j \in J} \lambda_{k+j} \left( \int_0^1 G_j(t, x, u) dt + g_j(x(1)) - \int_0^1 G_j[t] dt - g_j(x_*(1)) \right) \\ &\quad + \int_0^1 \langle p, \dot{x} - Ax - Bu - b \rangle dt - \int_0^1 \langle p, \dot{x}_* - Ax_* - Bu_* - b \rangle dt \\ &= \int_0^1 \left( \sum_{i \in I} \lambda_i F_i(t, x, u) + \sum_{j \in J} \lambda_{k+j} G_j(t, x, u) dt - \sum_{i \in I} \lambda_i F_i[t] - \sum_{j \in J} \lambda_{k+j} G_j[t] \right) dt \\ &\quad + \sum_{i \in I} \lambda_i f_i(x(1)) + \sum_{j \in J} \lambda_{k+j} g_j(x(1)) - \sum_{i \in I} \lambda_i f_i(x_*(1)) - \sum_{j \in J} \lambda_{k+j} g_j(x_*(1)) \\ &\quad - \int_0^1 (\langle \dot{p} + pA, x - x_* \rangle + \langle pB - \zeta, u - u_* \rangle) dt - \int_0^1 \langle \zeta, u - u_* \rangle dt \\ &\quad + \langle p(1), x(1) - x_*(1) \rangle - \langle p(0), x(0) - x_*(0) \rangle \\ &\geq 0. \end{aligned}$$

By Lemma 1,  $(x_*, u_*)$  is a weakly-efficient solution for  $(P)$ .  $\square$

Using Theorem 3 and Lemma 3, we can easily show that the conditions (18)-(21) in Theorem 3 are also necessary-sufficient for efficient solutions of  $(P^*)$  under the following Slater constraint qualifications (H6).

(H6): For every  $i \in I$ , there is an admissible process  $(x_i, u_i)$  for  $(P^*)$ , such that  $\mathcal{F}_j(x_i, u_i) - \mathcal{F}_i(x_*, u_*) < 0$  for any  $j \in I/\{i\}$  and  $\mathcal{G}_j(x_i, u_i) - \mathcal{G}_j(x_*, u_*) < 0$  for any  $j \in \{j \in J : \mathcal{G}_j(x_*, u_*) = 0\}$

**Theorem 4:** Assume that (H1)-(H6) and (A1) are satisfied. An admissible process  $(x_*, u_*)$  is an efficient solution for  $(P^*)$  if and only if there exist  $(\lambda_1 \dots, \lambda_{k+l}) \geq 0$  with  $(\lambda_1 \dots, \lambda_k) \gg 0$ ,  $p(\cdot) \in AC([0, 1], R^m)$ , and  $\zeta(\cdot) \in L^\infty([0, 1], R^n)$  such that (18)-(21) hold.

*Remark.* It is easy to see that the sufficiency in Theorem 3 and Theorem 4 also hold under the following simpler assumptions:  $F_i$  for  $i \in I$  and  $G_j$  for  $j \in I$  are convex in  $(x, u)$  and measurable in  $t$ ,  $f_i$  for  $i \in I$  and  $g_j$  for  $j \in I$  are convex functions,  $C$  is convex set and  $U(t)$  is convex a.e..

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